

Tagoram: Real-Time Tracking of Mobile RFID Tags to High-Precision Accuracy Using COTS Devices

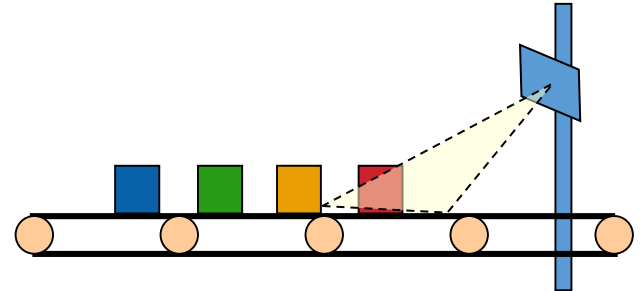
 Xiang-Yang Li

Lei Yang, Yekui Chen, Xiang-Yang Li,
Chaowei Xiao, Mo Li, Yunhao Liu

Sep. 9, 2014



清华大学
Tsinghua University



Tagoram: Real-Time Tracking of Mobile RFID Tags to High-Precision Accuracy Using COTS Devices



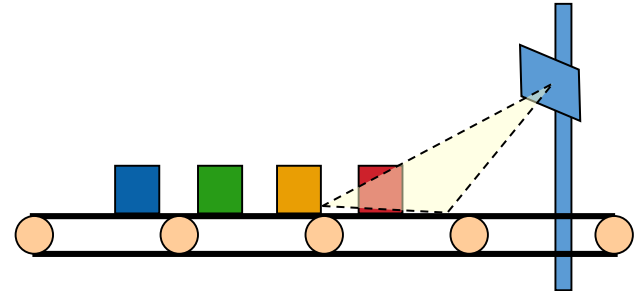
Xiang-Yang Li

Lei Yang, Yekui Chen, Xiang-Yang Li,
Chaowei Xiao, Mo Li, Yunhao Liu

Sep. 9, 2014



清华大学
Tsinghua University



Outline

01. Motivation

02. State-of-the-art

03. Overview

04. Movement with known track

05. Movement with unknown track

06. Implementation & Evaluation

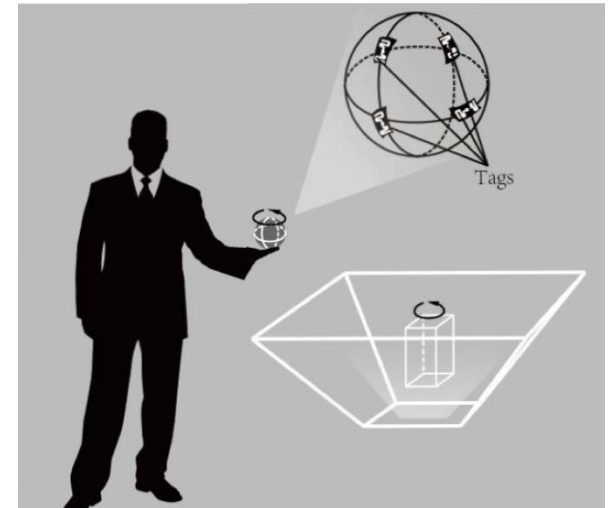
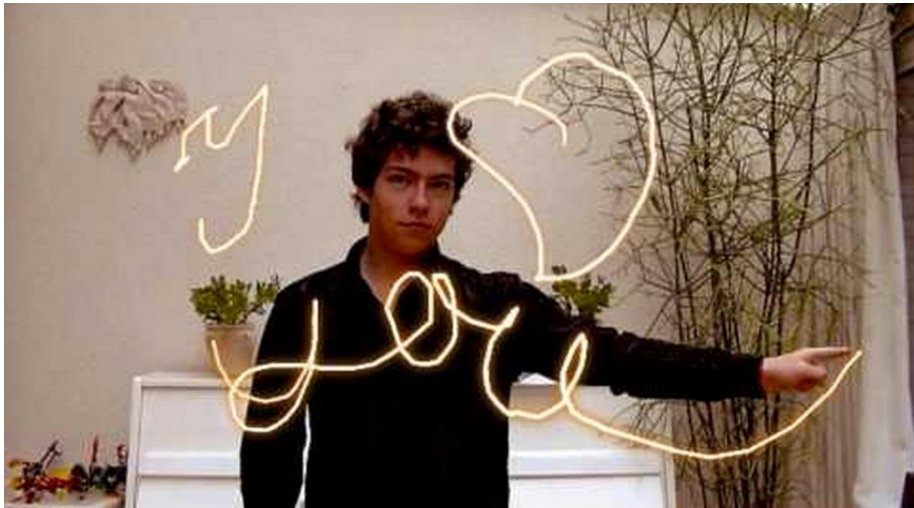
07. Pilot Study

08. Conclusion



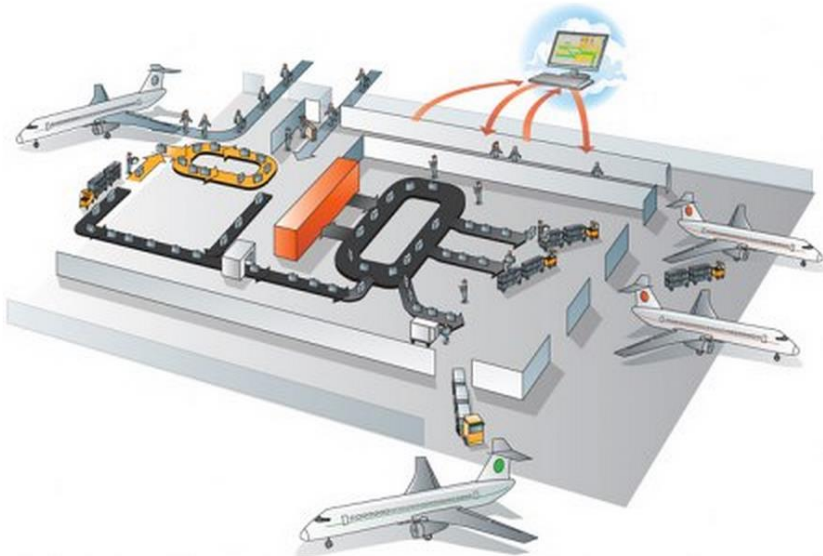
1 Motivation

**Imagine you can localize RFIDs to within
0.1cm to 1 cm!**



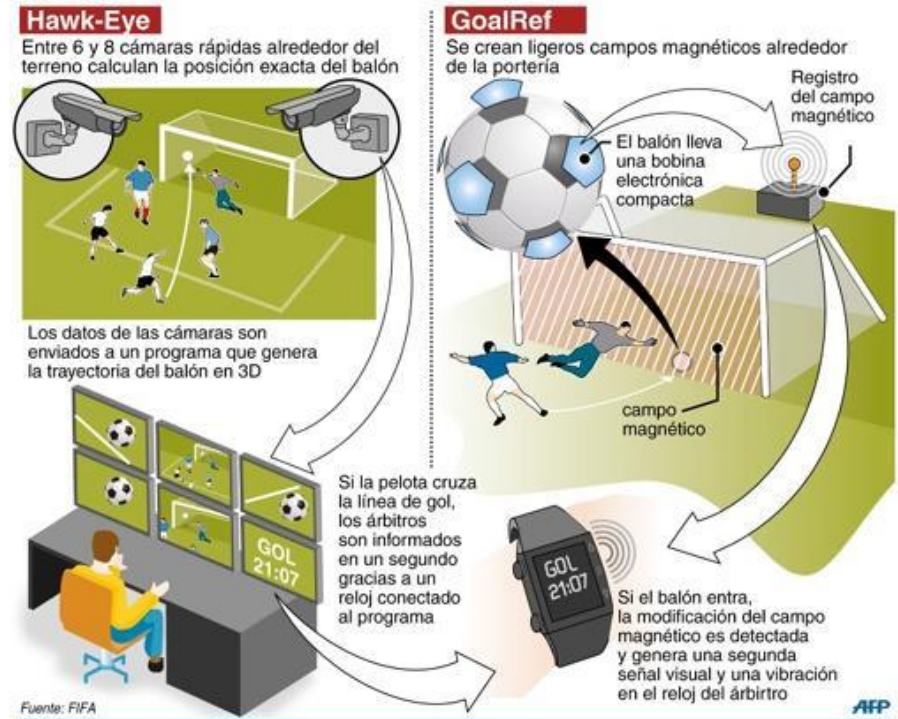
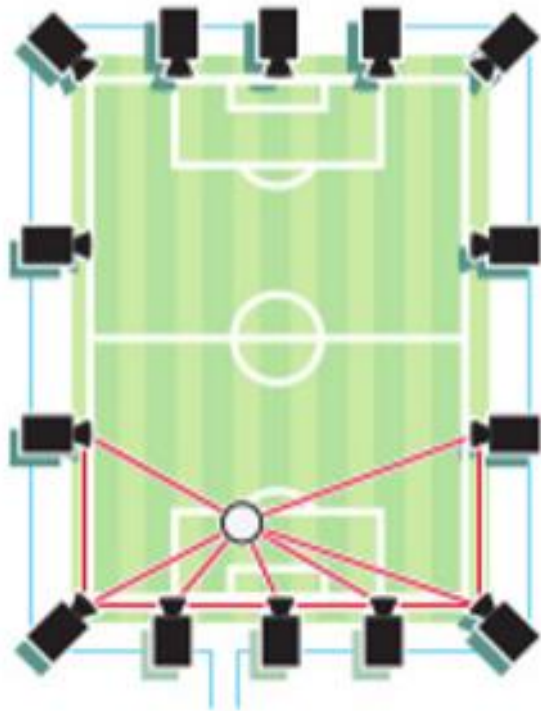
Human-computer interface

**Imagine you can localize RFIDs to within
0.1cm to 1 cm!**



Baggage sortation in airport

Imagine you can localize RFIDs to within 0.1cm to 1 cm!



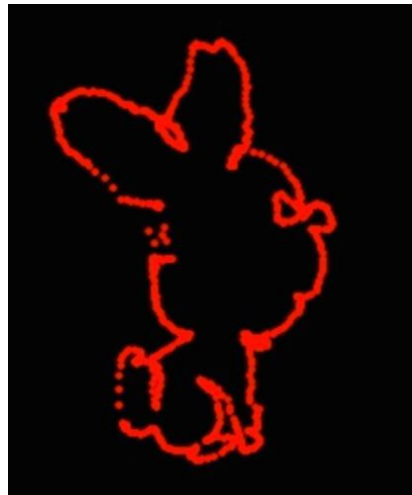
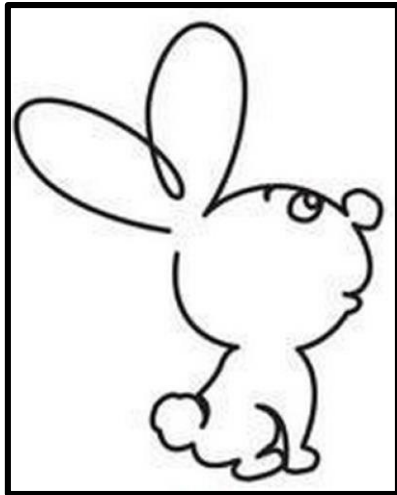
Goal-line technology

Demonstration

[http://young.tagsys.org/tracking/t
agoram/youtube](http://young.tagsys.org/tracking/agoram/youtube)

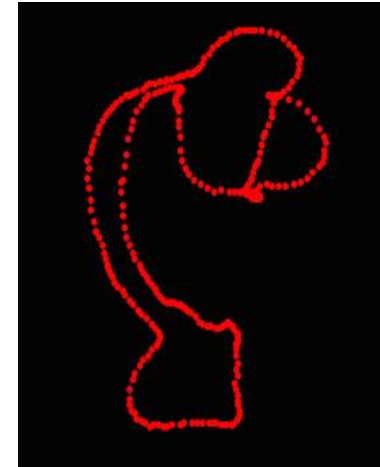
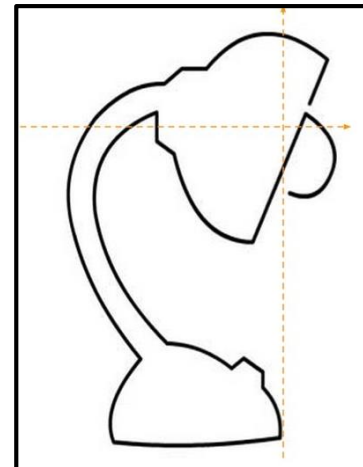
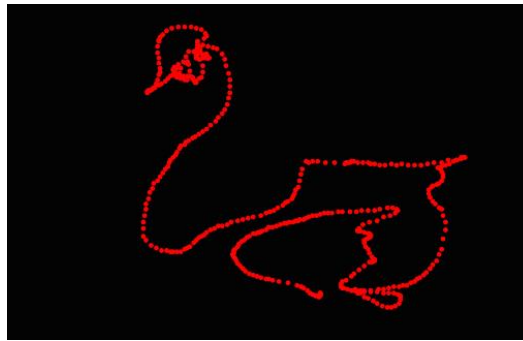
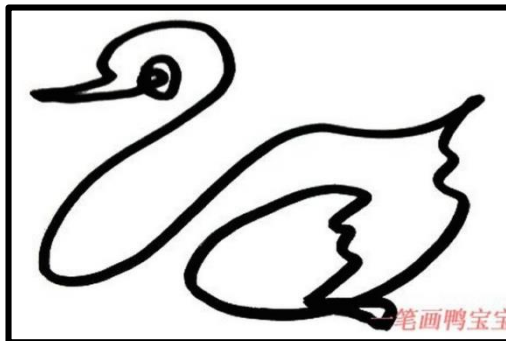
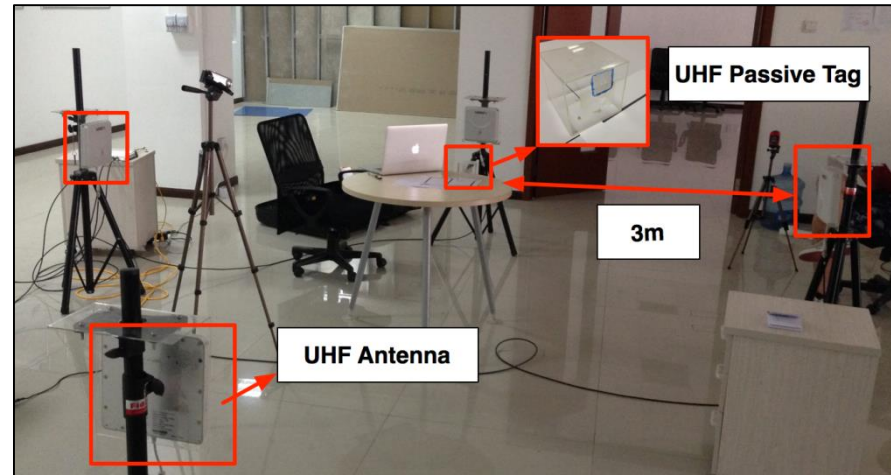
High-Precision RFID Tracking Using COTS Devices

Drawing in the Air

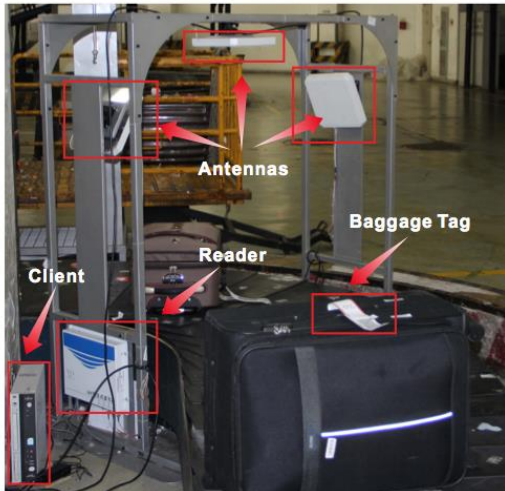


40cm

30cm



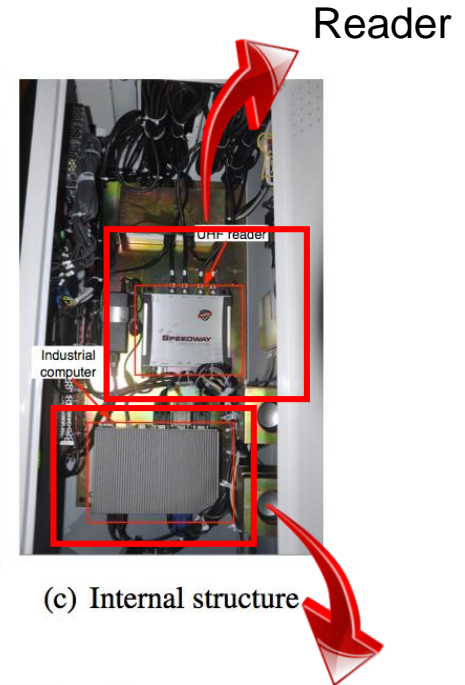
TrackPoint Deployed at Airports



(a) Version 1.0



(b) Version 2.0



(c) Internal structure



(a) Two TrackPoints



(b) Version 2.0

Industrial
computer



2 State-of-the-art

RFID Tracking & Localization

State-of-the-art

LANDMARC

Lionel M. Ni

1120mm

2004

2010

2011

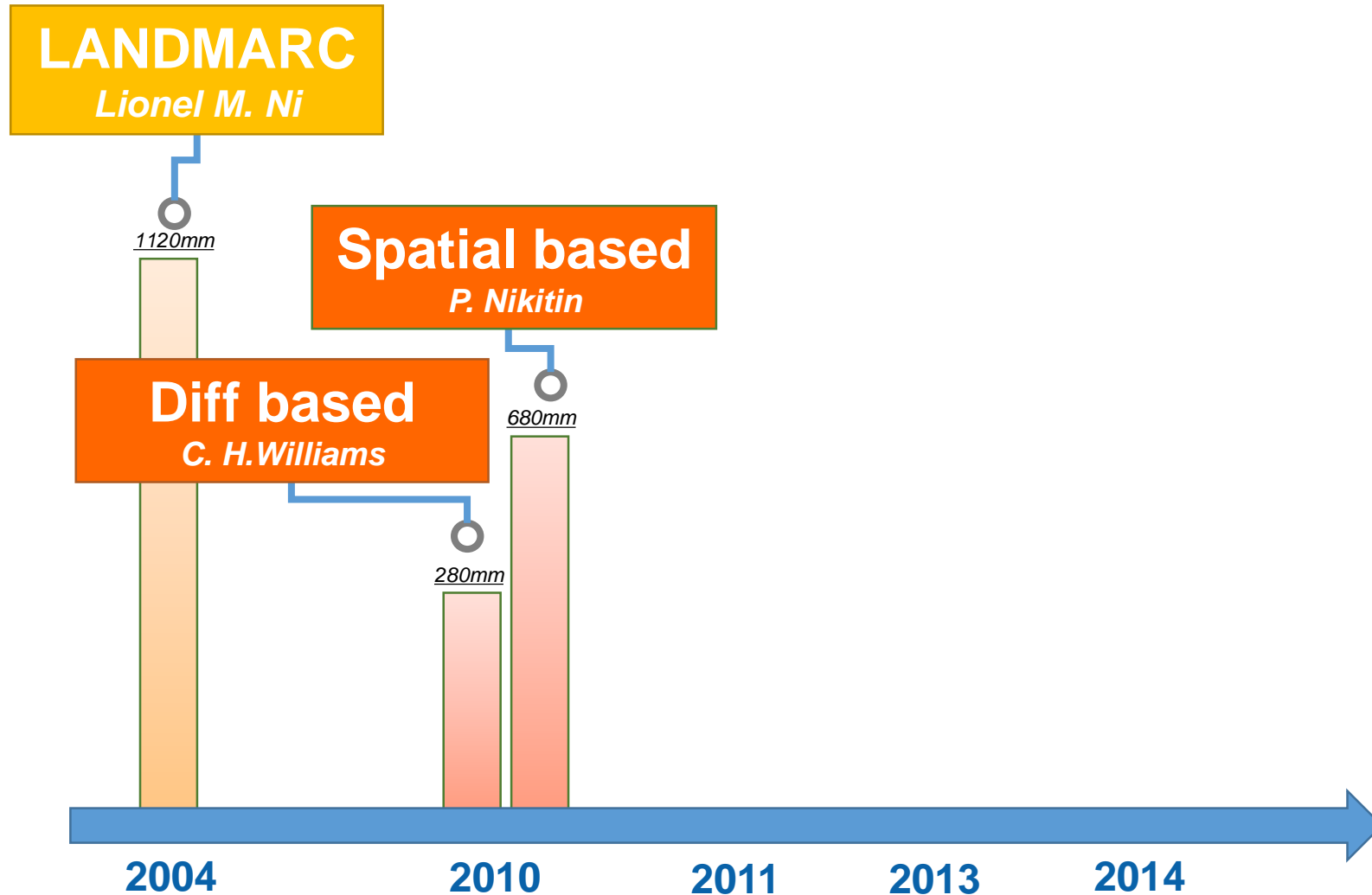
2013

2014



RFID Tracking & Localization

State-of-the-art



RFID Tracking & Localization State-of-the-art

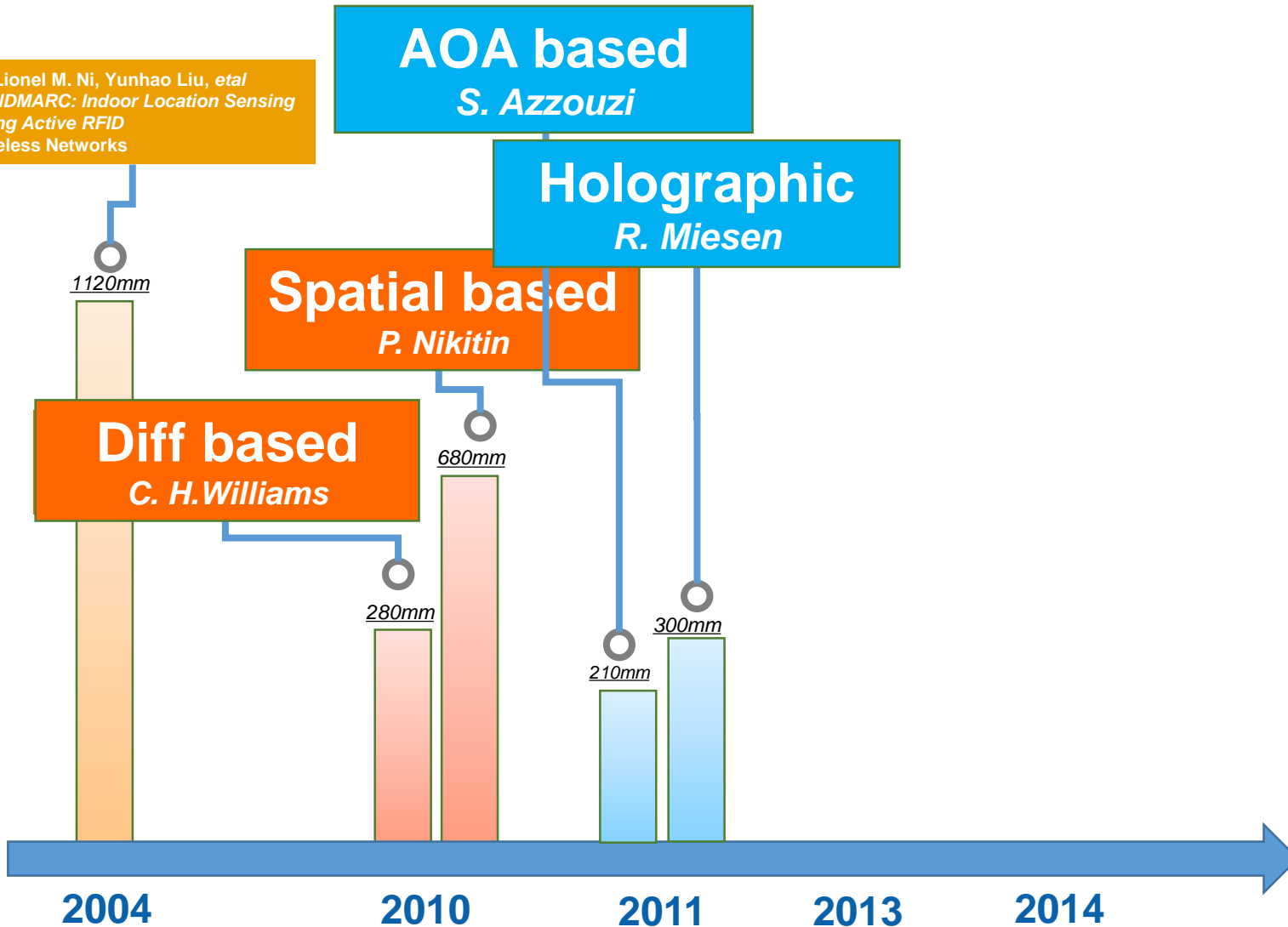
[1] Lionel M. Ni, Yunhao Liu, et al
*LANDMARC: Indoor Location Sensing
Using Active RFID
Wireless Networks*

AOA based
S. Azzouzi

Holographic
R. Miesen

Spatial based
P. Nikitin

Diff based
C. H. Williams



RFID Tracking & Localization

State-of-the-art

[1] Lionel M. Ni, Yunhao Liu, et al
LANDMARC: Indoor Location Sensing Using Active RFID Wireless Networks

AOA based
S. Azzouzi

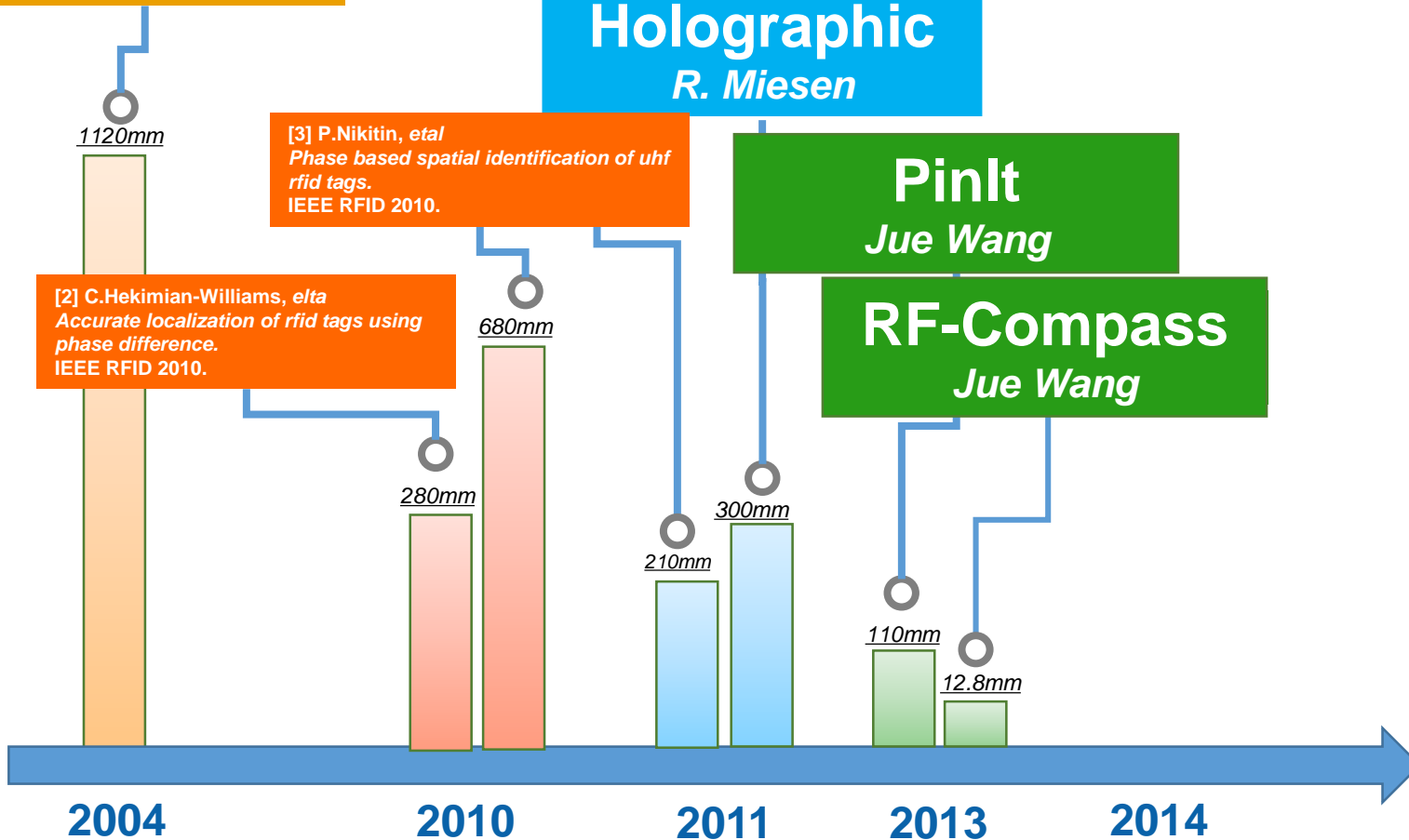
Holographic
R. Miesen

[3] P.Nikitin, et al
Phase based spatial identification of uhf rfid tags.
IEEE RFID 2010.

[2] C.Hekimian-Williams, et al
Accurate localization of rfid tags using phase difference.
IEEE RFID 2010.

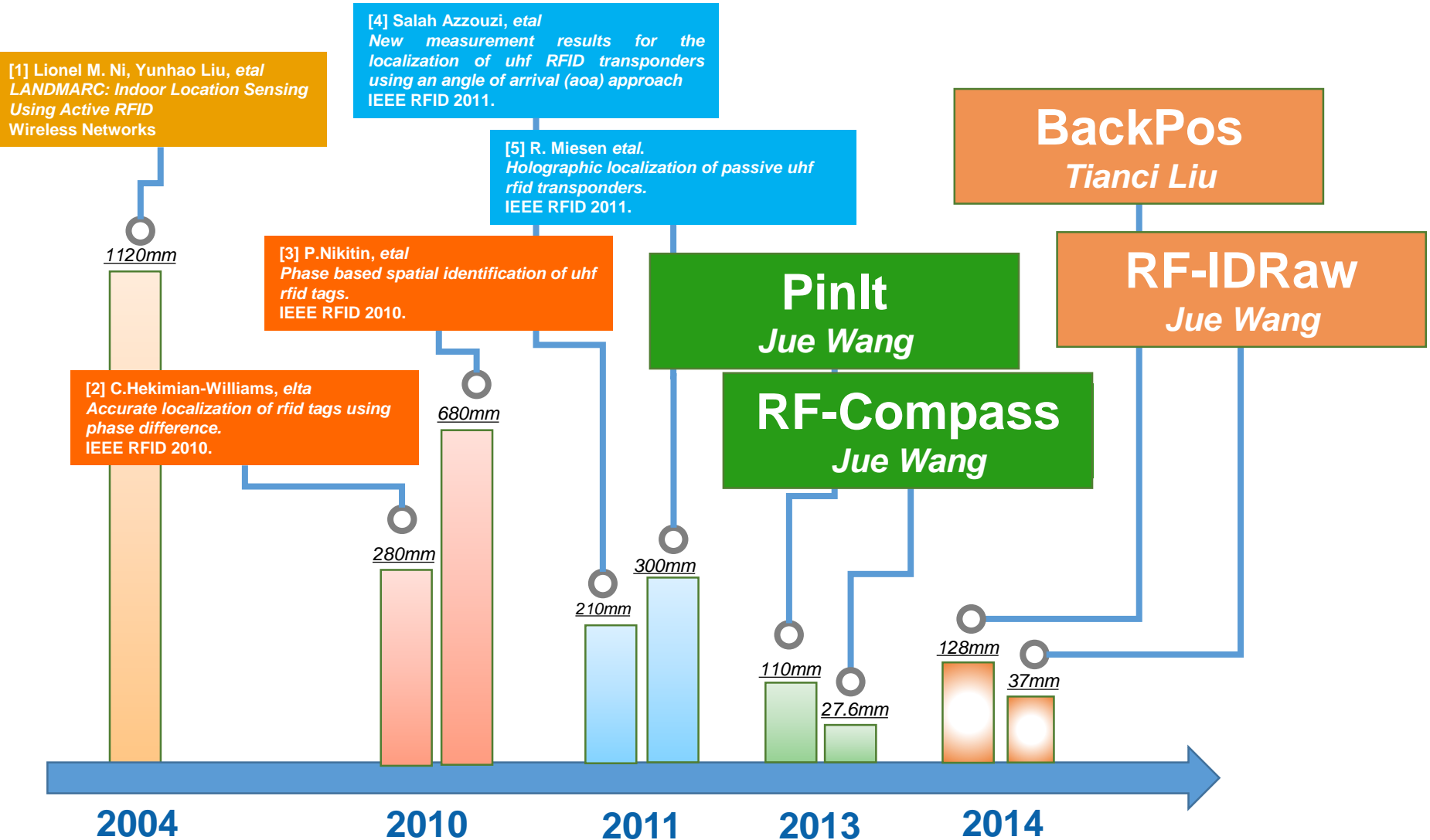
PinIt
Jue Wang

RF-Compass
Jue Wang



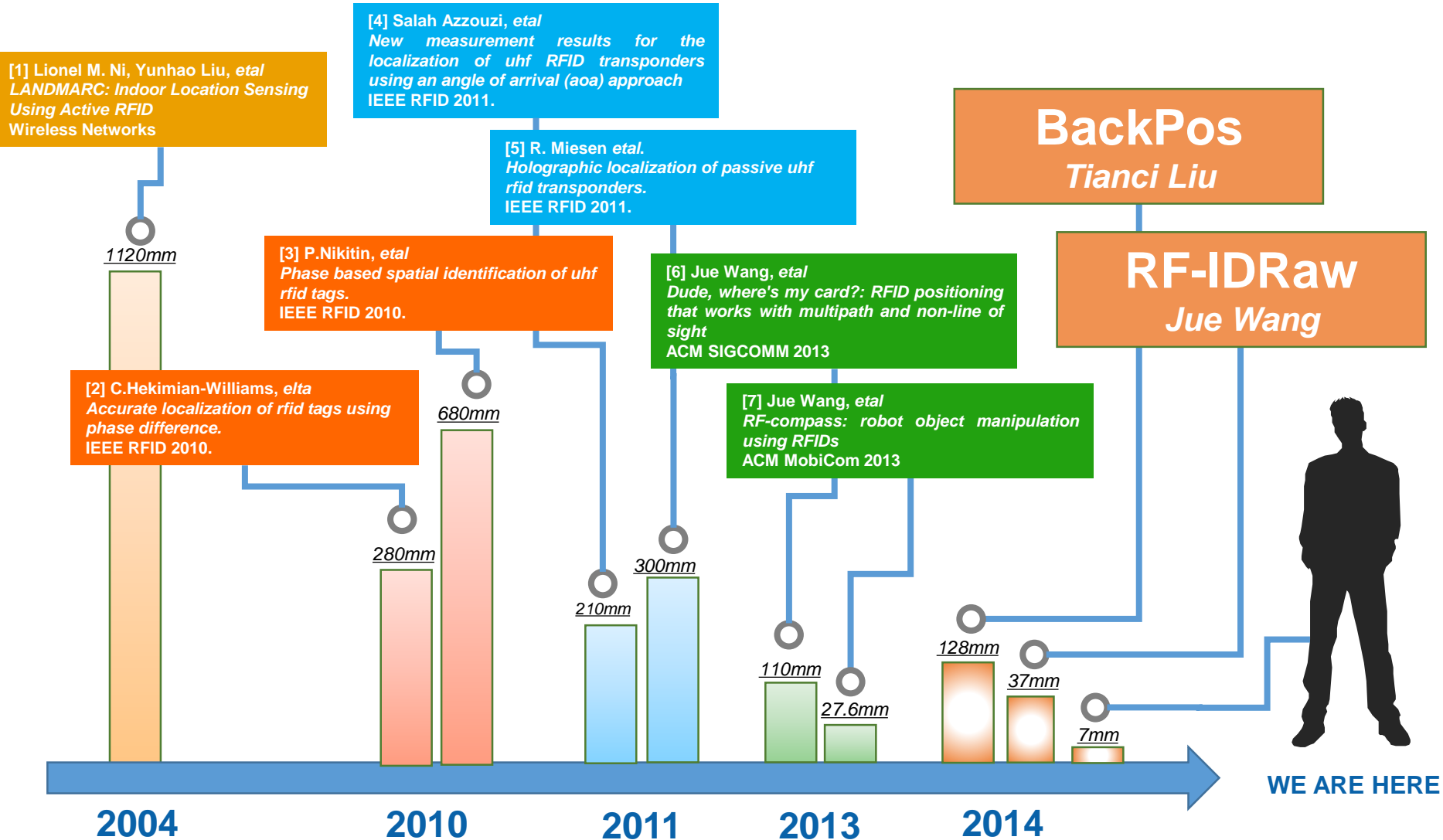
RFID Tracking & Localization

State-of-the-art



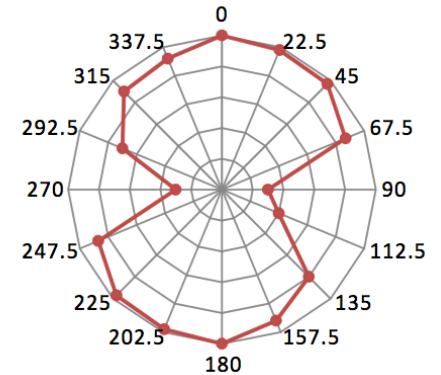
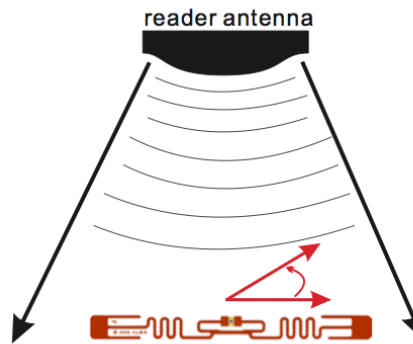
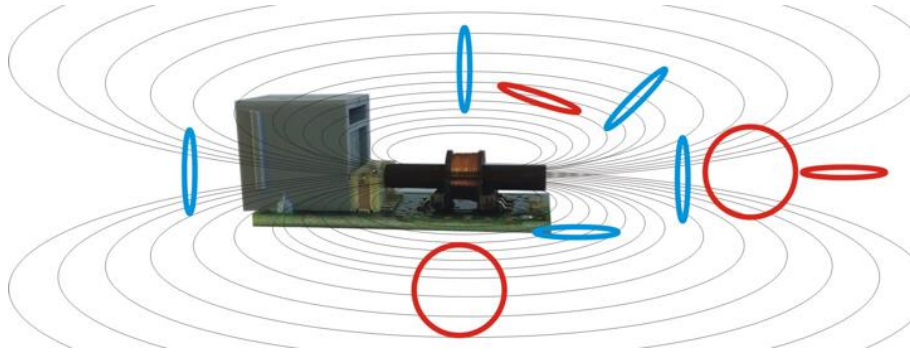
RFID Tracking & Localization

State-of-the-art



State-of-the-art Techniques

① RSS based Methods



Orientation

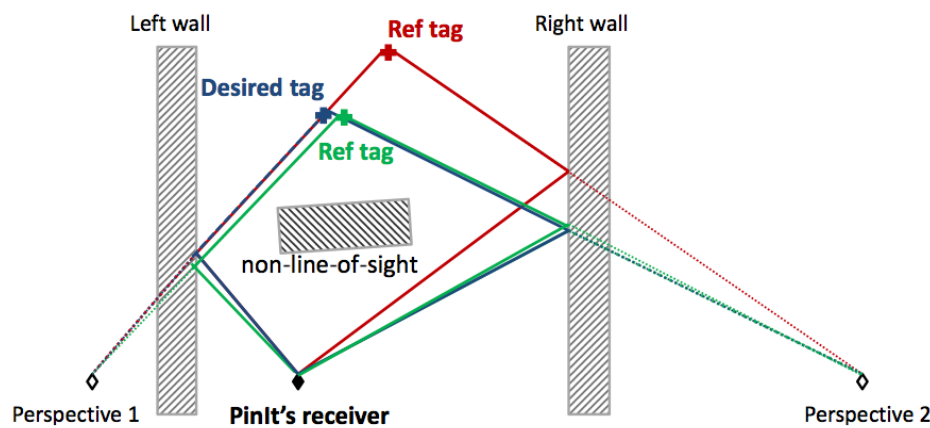
VS

RSS

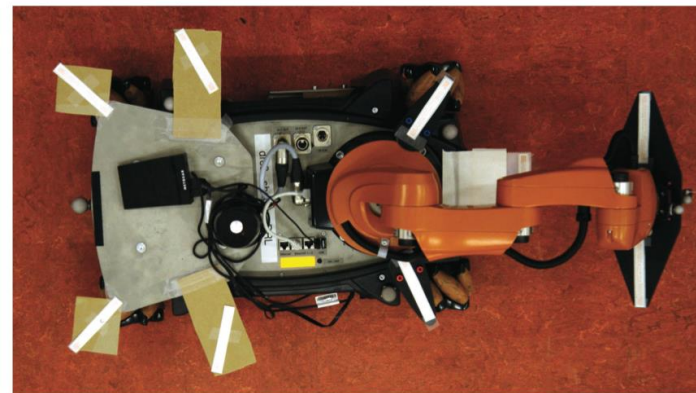
**RSS is not a reliable location indicator
especially for UHF tags**

State-of-the-art Techniques

2 Phase based Methods



PinIt (SIGCOMM 2013)

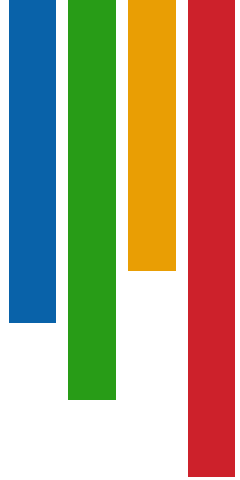


RF-Compass (MobiCom 2013)

Needs to deploy dense reference tags

Summary of Challenges

- **Need mm-level localization accuracy achieved**
 - especially for mobile tags.
- **Small overhead, COTS devices**
 - infeasible for using many references for a tracking system spanning a long pipeline.
- **Fast-changing environment**
 - multipath reflection of RF signals
 - varied orientation of tags
 - Doppler effect



3 How Tagoram works?

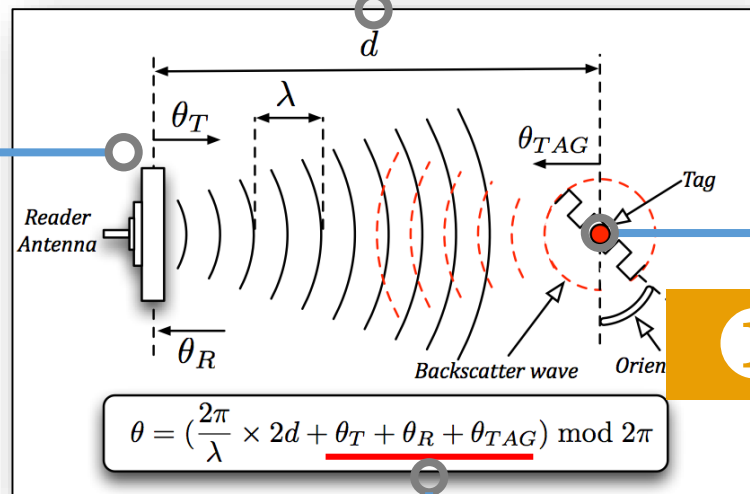
Overview the basic idea

Backscatter Communication

② *Double distance*

③ *Continuous wave*

① *Battery free*



④ *Device diversity*

COTS RFID Reader

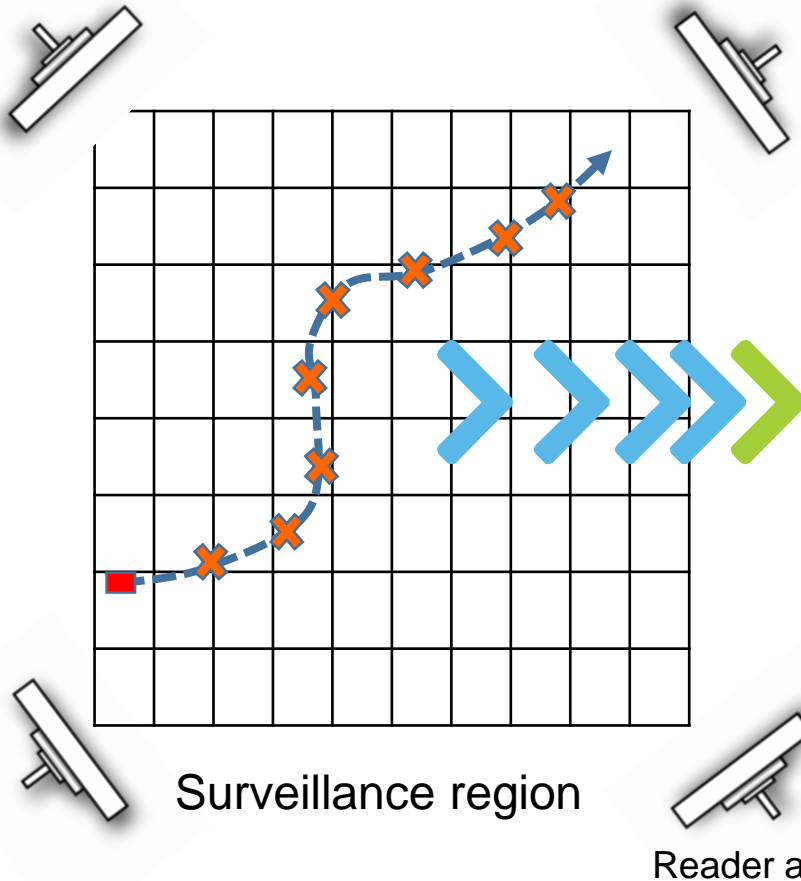
0.0015 radians
(4096 bits)

$\approx 0.038mm$
accuracy



Impinj Reader

Problem definition



$$M \times N$$

$$\begin{aligned} & \{(\theta_{1,1}, t_{1,1}), (\theta_{1,2}, t_{1,2}) \cdots, (\theta_{1,N}, t_{1,N})\} \\ & \{(\theta_{2,1}, t_{2,1}), (\theta_{2,2}, t_{2,2}) \cdots, (\theta_{2,N}, t_{2,N})\} \\ & \quad \vdots \quad \quad \quad \vdots \quad \quad \quad \vdots \\ & \{(\theta_{M,1}, t_{M,1}), (\theta_{M,2}, t_{M,2}) \cdots, (\theta_{M,N}, t_{M,N})\} \end{aligned}$$

Utilizing antennas' locations, sampled phase values and timestamps to find out the tag's trajectory $f(t)$?

Tagoram

Case 1. Controllable Case

Case 2. Uncontrollable Case



4

Movement with Known Track

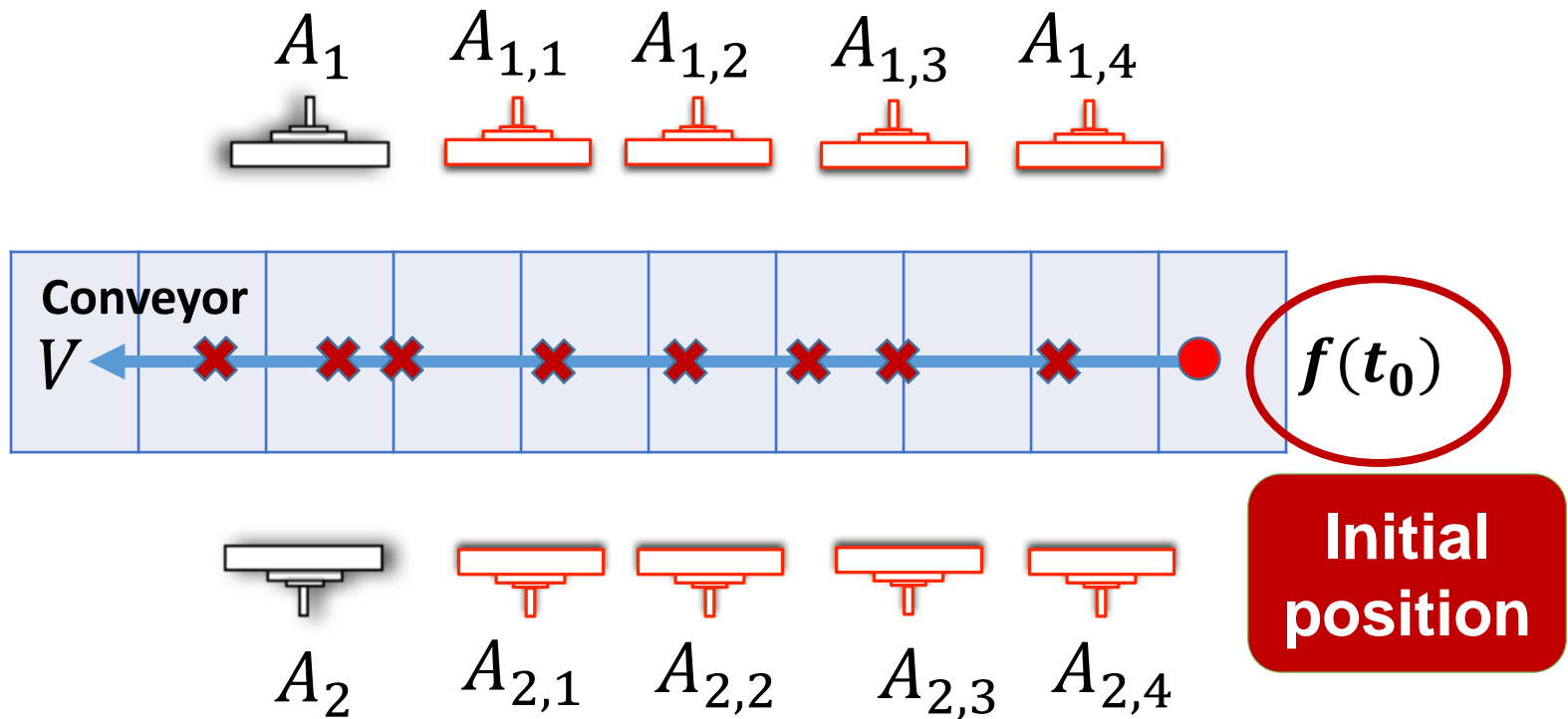
Controllable case



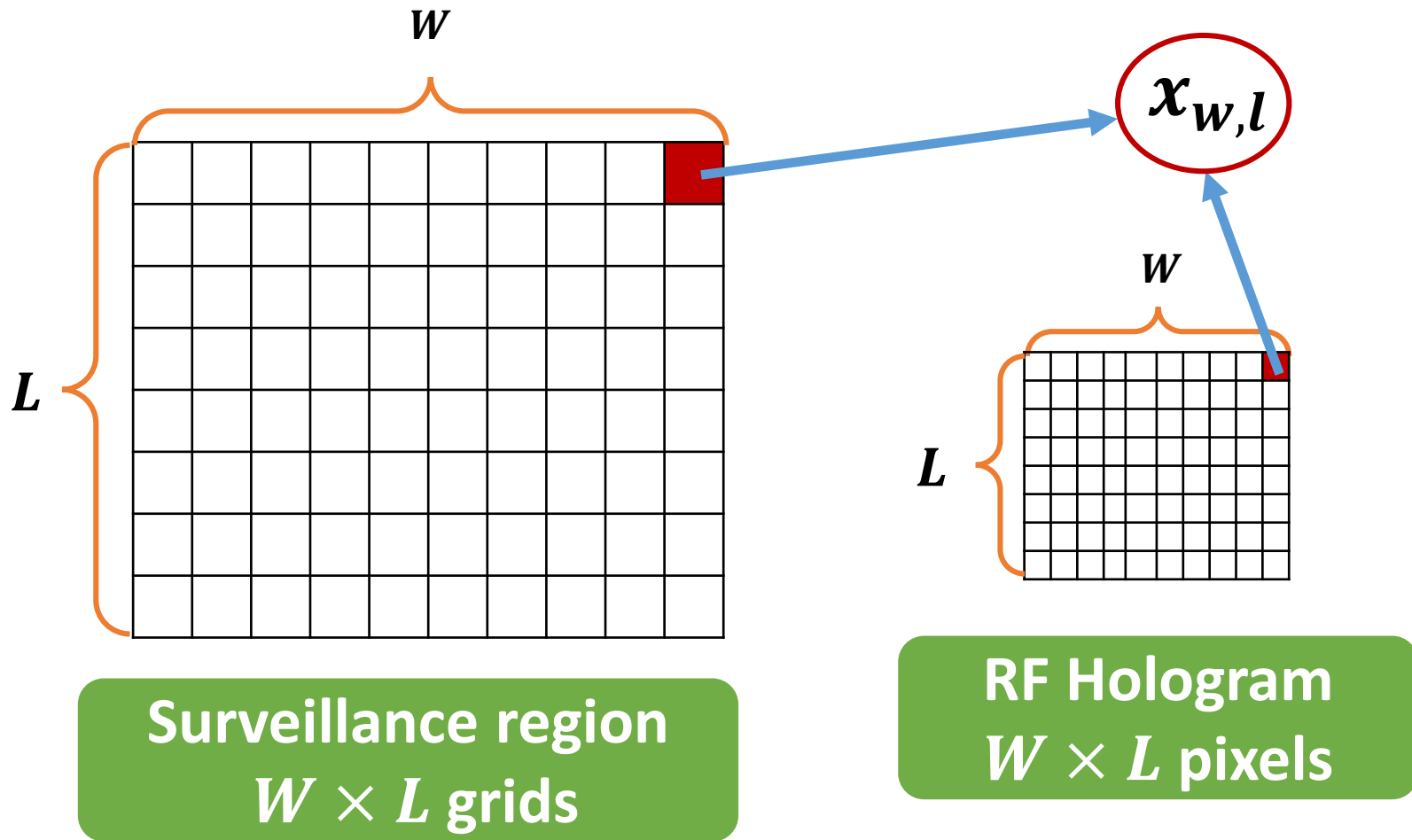
Our goal is to find the
tag's trajectory with
a known track.

Virtual Antenna Matrix

Inverse Synthetic Aperture Radar



RF Hologram



The key is

**How to define the
likelihood?**

RF Hologram

**Naïve
Hologram**



**Augmented
Hologram**



Thermal noise

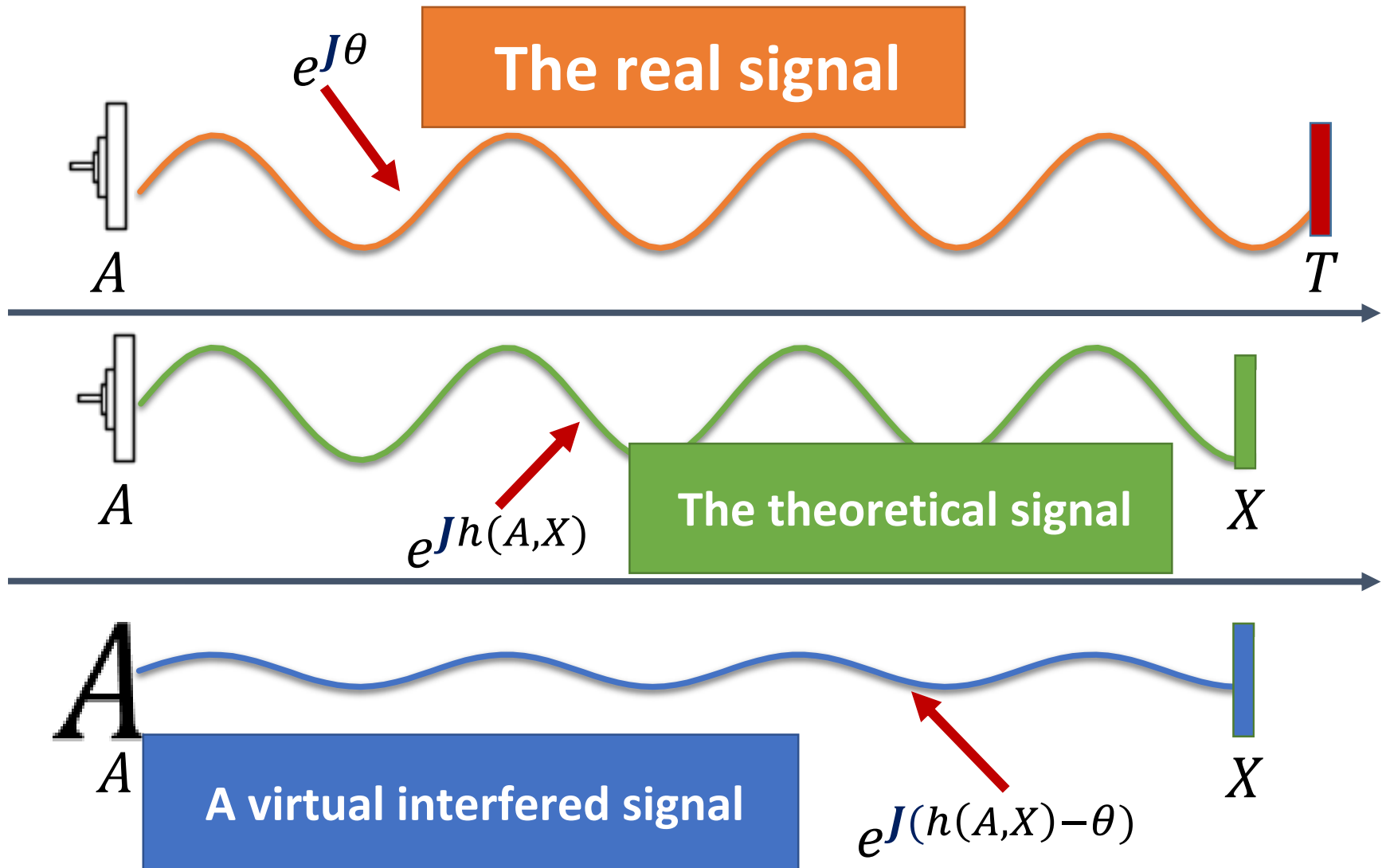


**Differential
Augmented
Hologram**



Device diversity

Naïve Hologram



Naïve Hologram

Sum of all signals

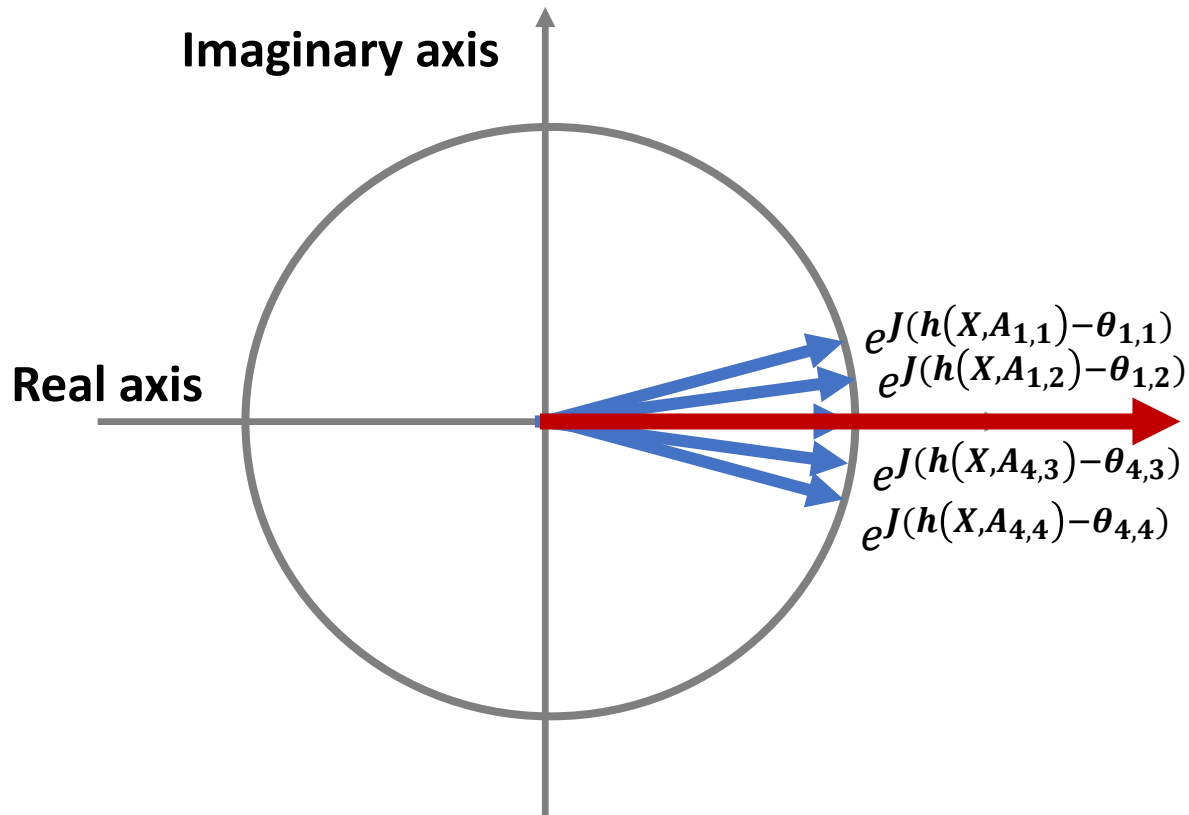
DEFINITION 1. The naive hologram is an image in which the pixel value $x_{w,l}$, indicating the likelihood that the corresponding grid $X_{w,i}$ is the initial position, is calculated by

$$x_{w,l} = \left| \sum_{m=1}^M \sum_{n=1}^N S(X_{w,l}, A_{m,n}, \theta_{m,n}) \right| \quad (9)$$

where $S(X, A, \theta) = e^{\mathbf{J}(h(X,A) - \theta)}$. The term \mathbf{J} denotes the imaginary number and the term $e^{\mathbf{J}\theta}$ represents a complex exponential

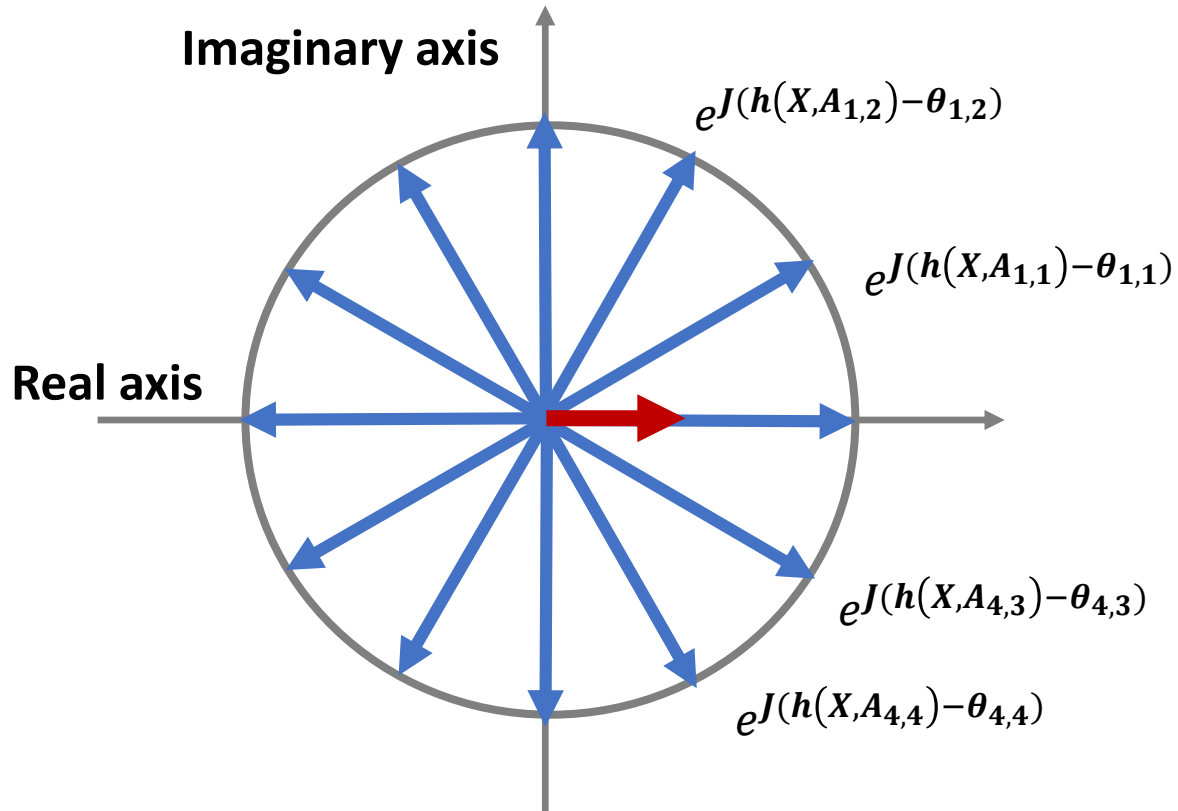
Virtual interfered signal

Naïve Hologram



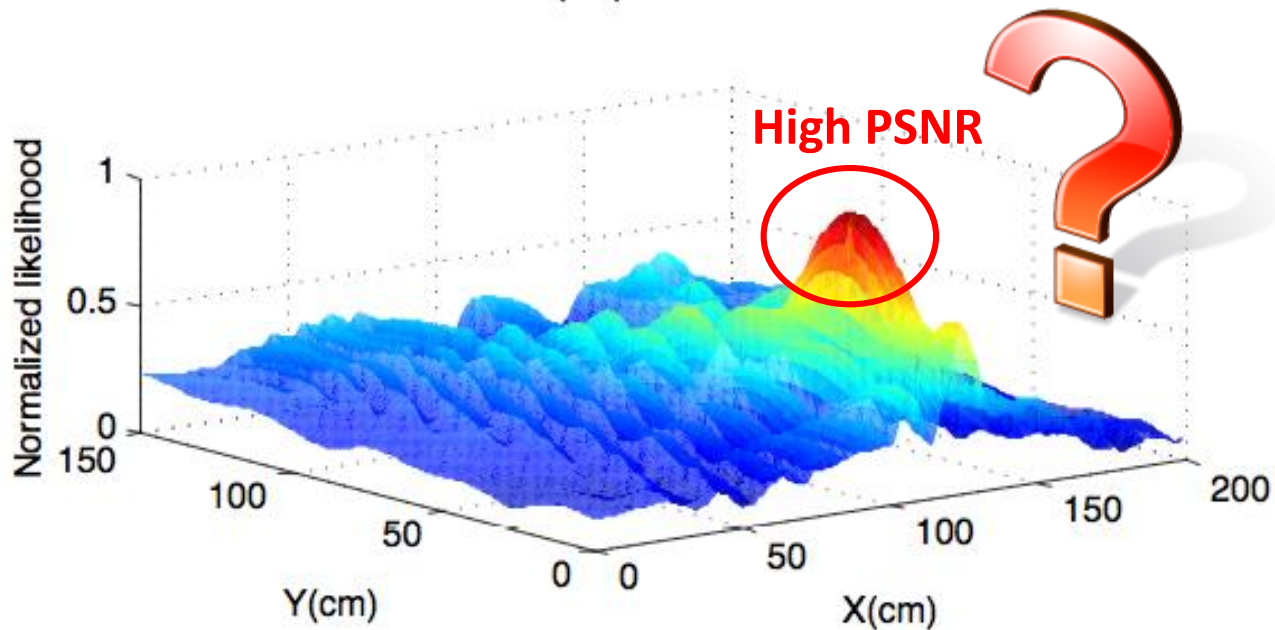
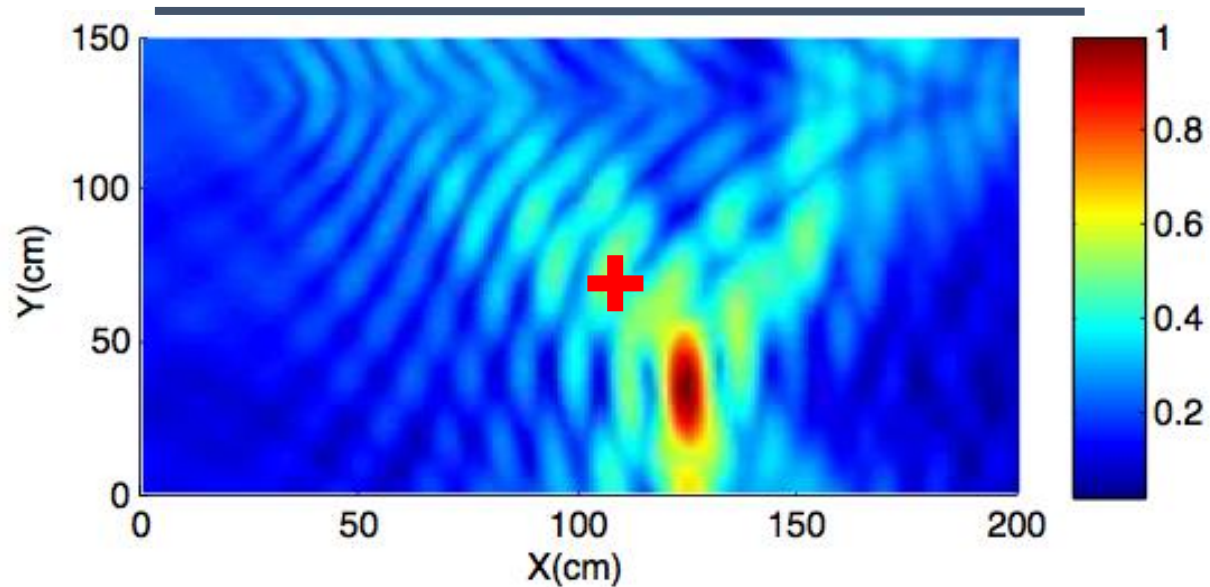
If X is the initial location, the waves add up constructively.

Naïve Hologram



If X is not the initial location, the waves canceled out.

Naïve Hologram



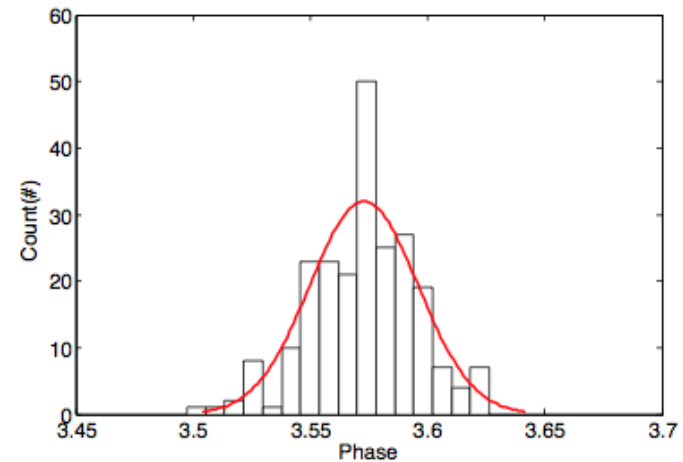
Influence from Thermal Noise

100 tags

$0^\circ \sim 40^\circ$

920 ~ 926 MHz

-70 ~ -30dbm



Experiment

Observations

Modeling

CDF

$$\theta \sim \mathcal{N}(\mu, \sigma)$$

$$F(\theta; \mu, \sigma)$$

varying with d

$$\sigma = 0.1$$

**How to deal with
thermal noise?**

Augmented Hologram

Augmented Hologram

A probabilistic weight

DEFINITION 2 (AH). The augmented hologram is an image in which the pixel value $x_{w,l}$ is

$$(h(X, A) - \theta) \sim \mathcal{N}(0, 0.1)$$

$$x_{w,l} = \left| \sum_{m=1}^M \sum_{n=1}^N \|S(X_{w,l}, A_{m,n}, \theta_{m,n})\| S(X_{w,l}, A_{m,n}, \theta_{m,n}) \right| \quad (10)$$

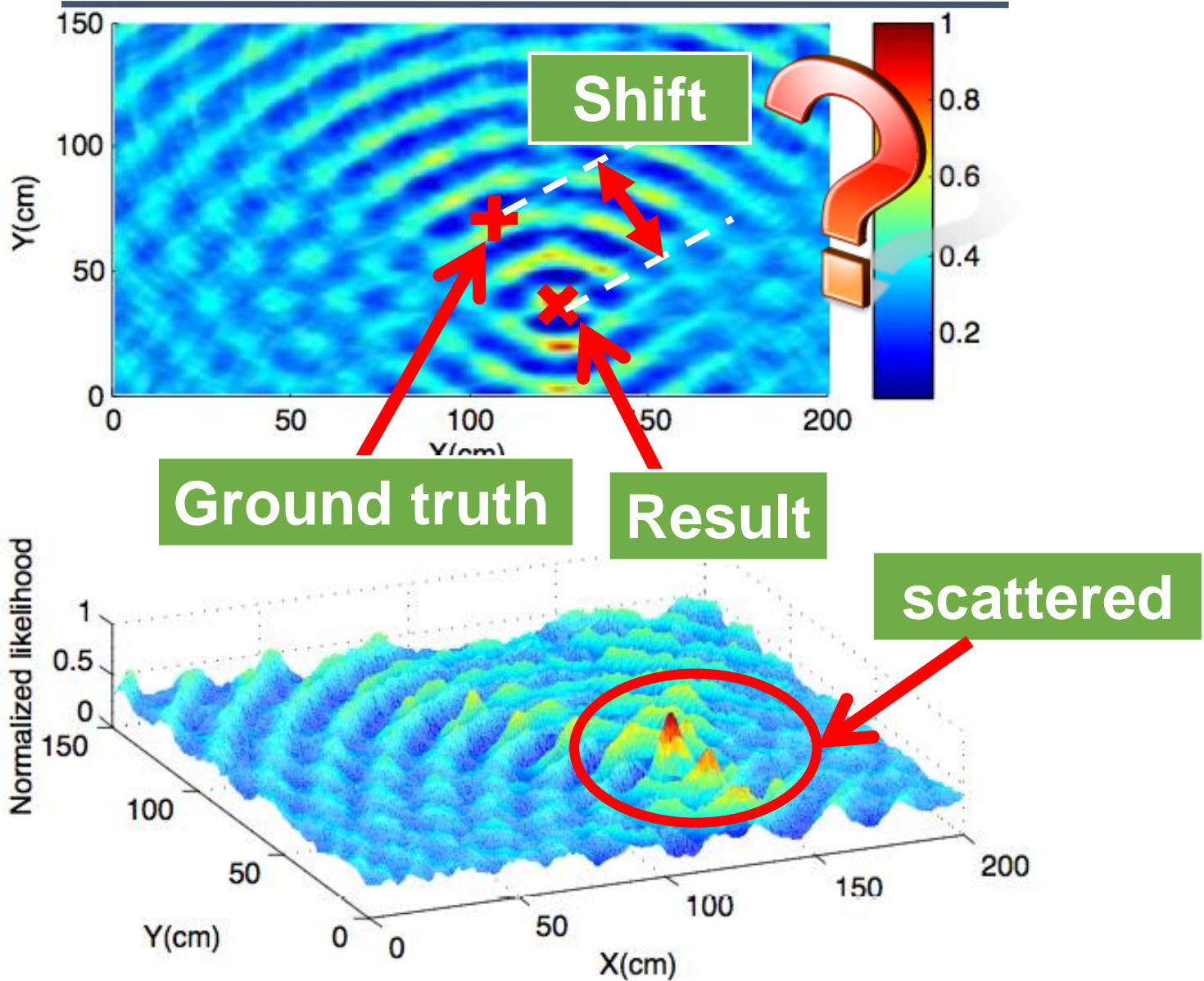
where

$$\begin{cases} \|S(X, A, \theta)\| = 2 \times F(|h(X, A) - \theta|; 0, 0.1) \\ F(x; \mu, \sigma) = \frac{1}{\sigma\sqrt{2\pi}} \int_x^\infty \exp\left(-\frac{(t - \mu)^2}{2\sigma^2}\right) dt \end{cases}$$

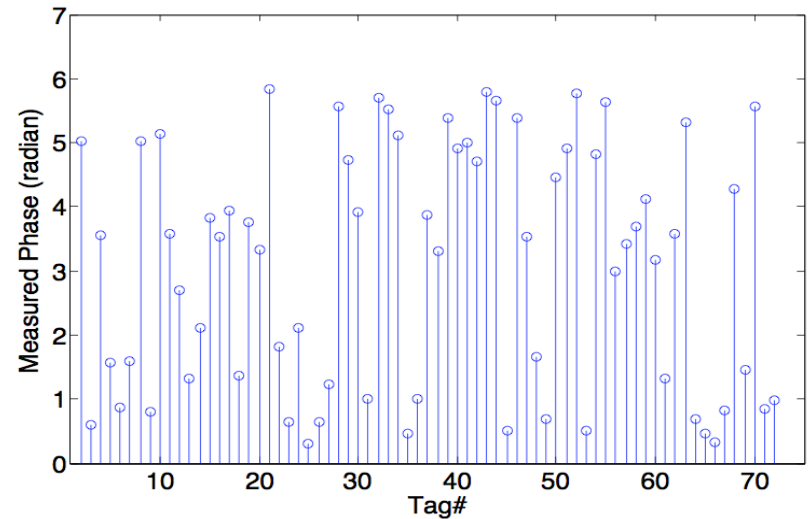
and $F(x; \mu, \sigma)$ is the cumulative probability function of Gaussian distribution $\mathcal{N}(\mu, \sigma)$.

Probability of $T \rightarrow A$

Augmented Hologram



Influence from Tag Diversity



Experiment

Observations

Modeling

Random test

$$\begin{cases} \theta = \left(\frac{2\pi}{\lambda} \times 2d + c\right) \bmod 2\pi \\ c = \theta_T + \theta_R + \theta_{TAG} \end{cases}$$

Tag diversity



Pass ***KS-test*** to be verified over a ***uniform distribution*** with 0.5 significant level.

How to eliminate tag diversity?

Differential Augmented
Hologram

Differential Augmented Hologram

DEFINITION 3 (DAH). *The differential augmented hologram is an image in which the pixel value is calculated by*

$$x_{w,l} = \left| \sum_{m=1}^M \sum_{n=1}^N \|\mathbb{S}(X_{w,l}, A_{m,n}, \theta_{m,n})\| \mathbb{S}(X_{w,l}, A_{m,n}, \theta_{m,n}) \right| \quad (11)$$

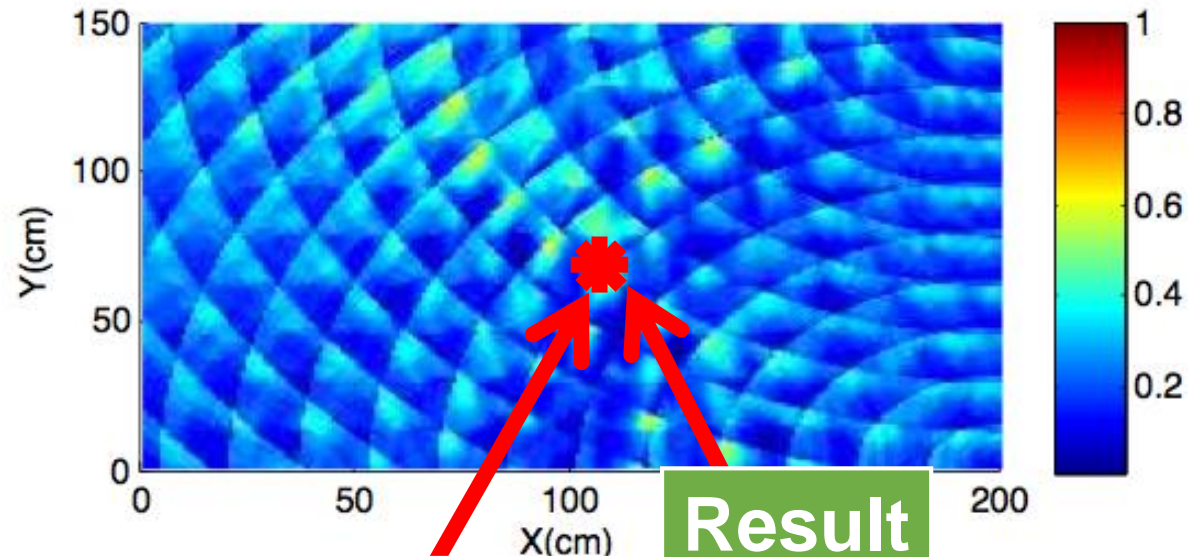
where

$$\begin{cases} \mathbb{S}(X_{w,l}, A_{m,n}, \theta_{m,n}) = e^{\mathbf{J}\theta_{dif}} \\ \|\mathbb{S}(X_{w,l}, A_{m,n}, \theta_{m,n})\| = 2 \times F(|\theta_{dif}|; 0, 0.1 \times \sqrt{2}) \\ \theta_{dif} = (h(X_{w,l}, A_{m,n}) - \theta_{m,n}) - (h(X_{w,l}, A_{m,1}) - \theta_{m,1}) \end{cases}$$



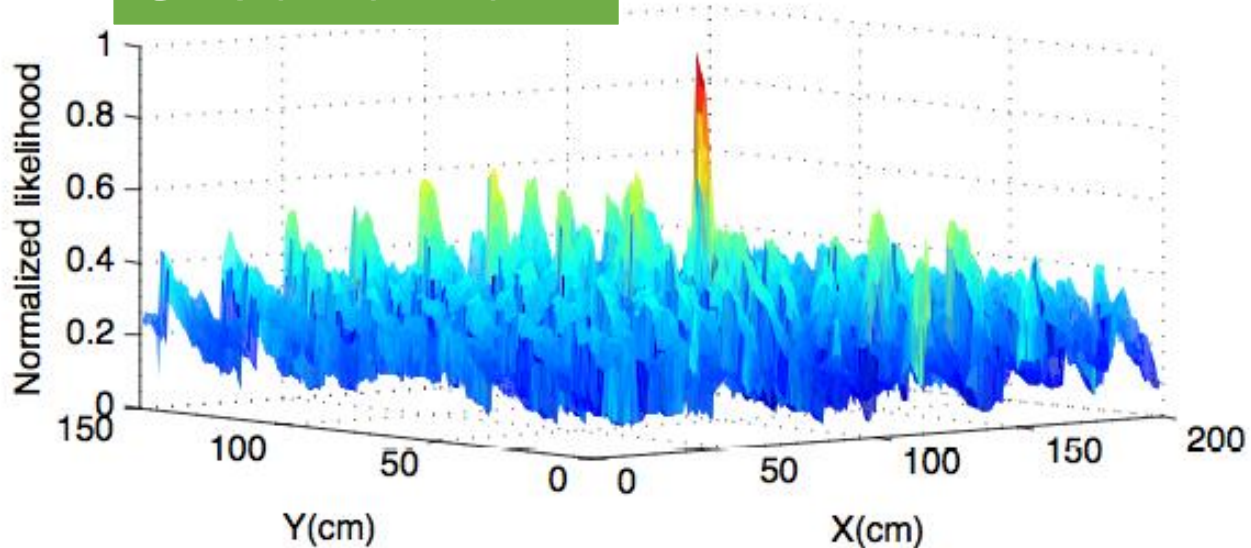
Using the difference of phase values

Differential Augmented Hologram



Result

Ground truth





5

Movement with Unknown Track

Uncontrollable case

Overview

- 1 Estimating Speeds,
Fitting tag's trajectory**
- 2 Selecting the optimal trajectory**



6

Implementation & Evaluation

Purely based on COTS devices

Hardware & Software Introduction

Reader

ImpinJ R420 reader.



4 directional antennas



Tag

Alien 2 × 2 Inlay



Alien Squiggle Inlay



Software

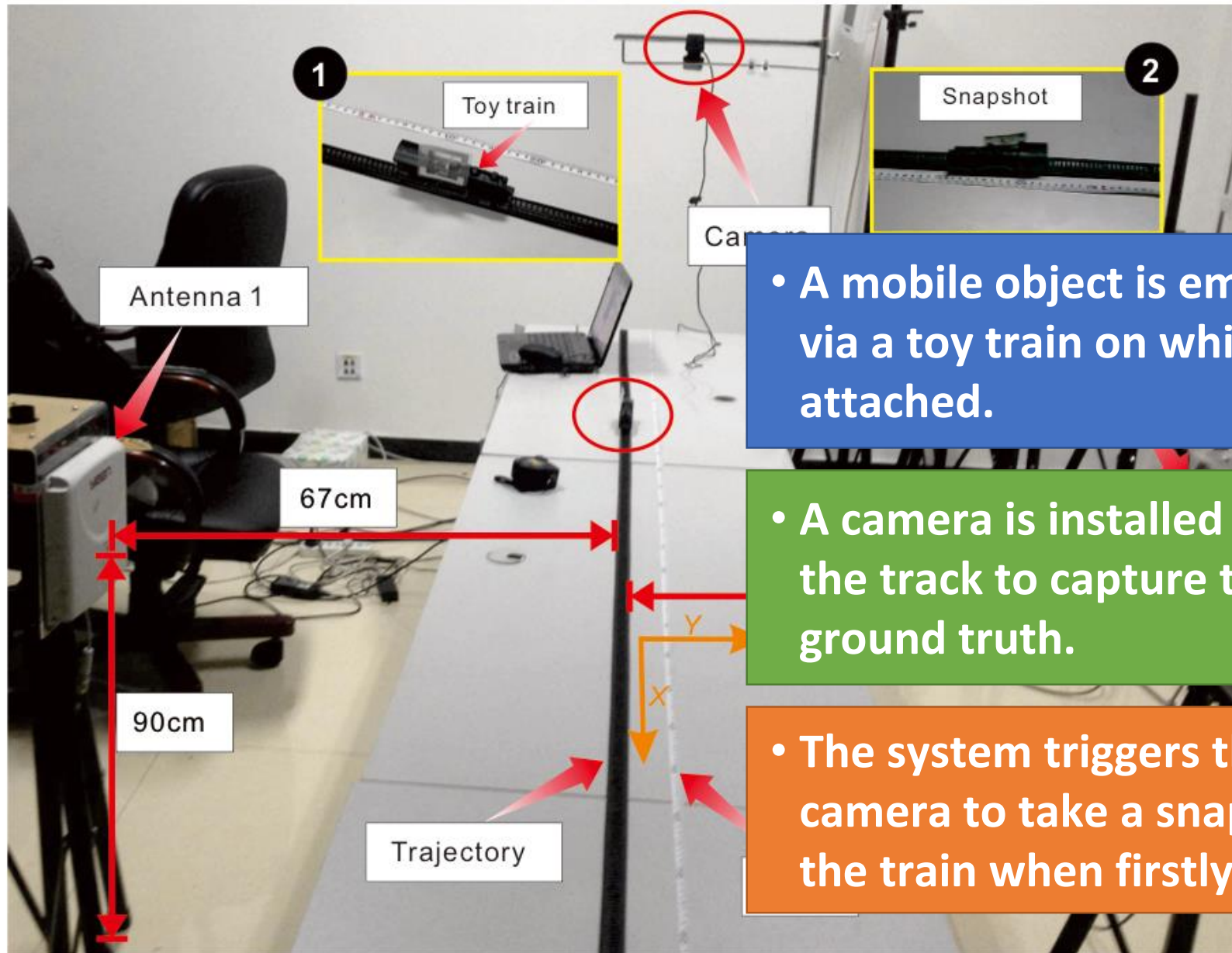
EPCglobal LLRP



Java



Linear Track

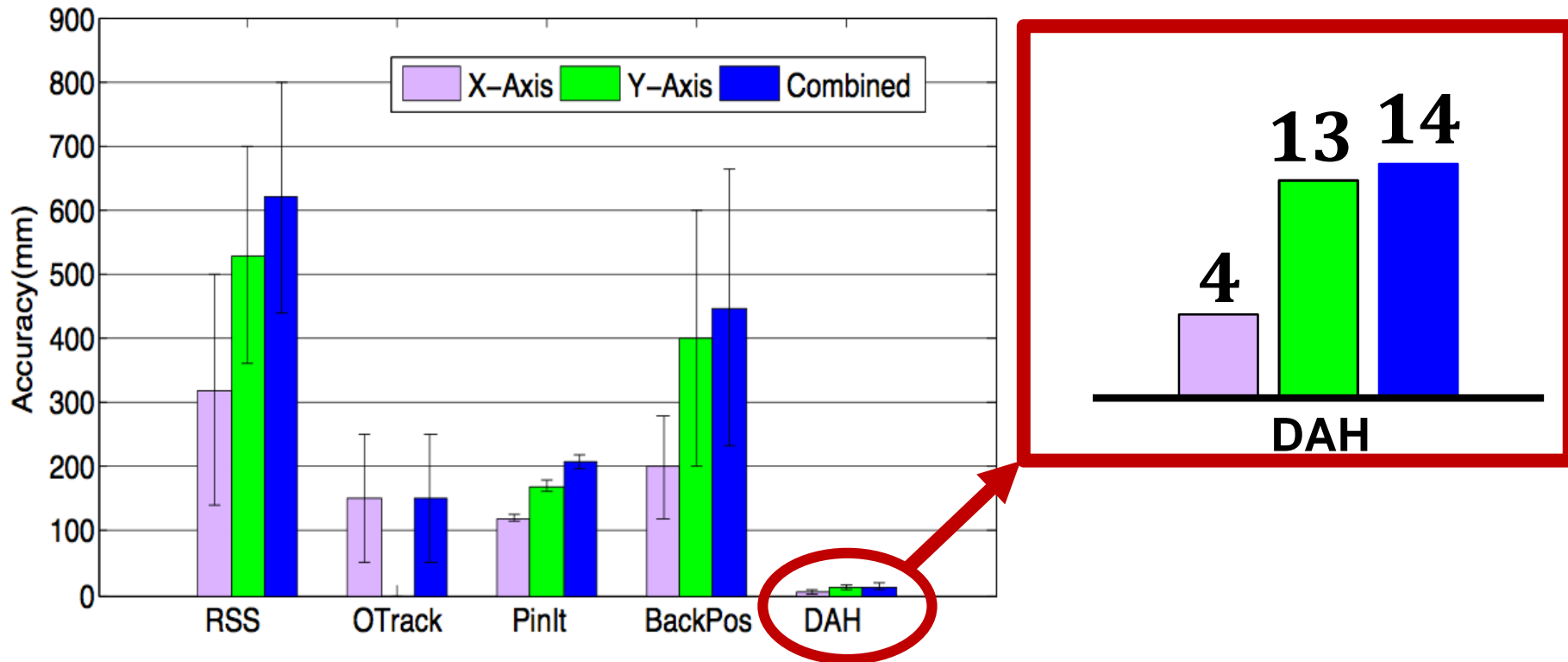


- A mobile object is emulated via a toy train on which a tag is attached.

- A camera is installed above the track to capture the ground truth.

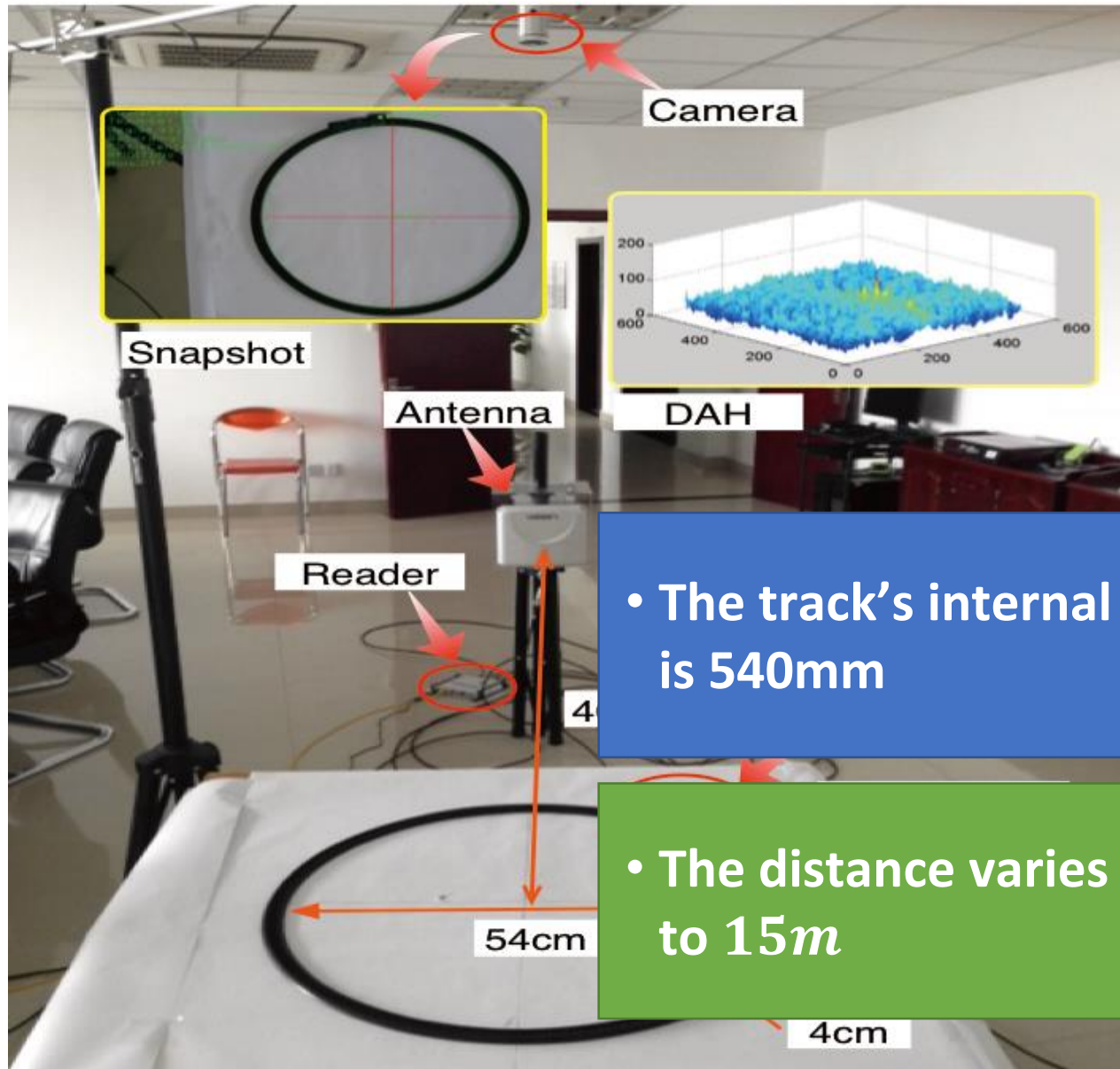
- The system triggers the camera to take a snapshot on the train when firstly read.

Tracking Accuracy in Linear Track

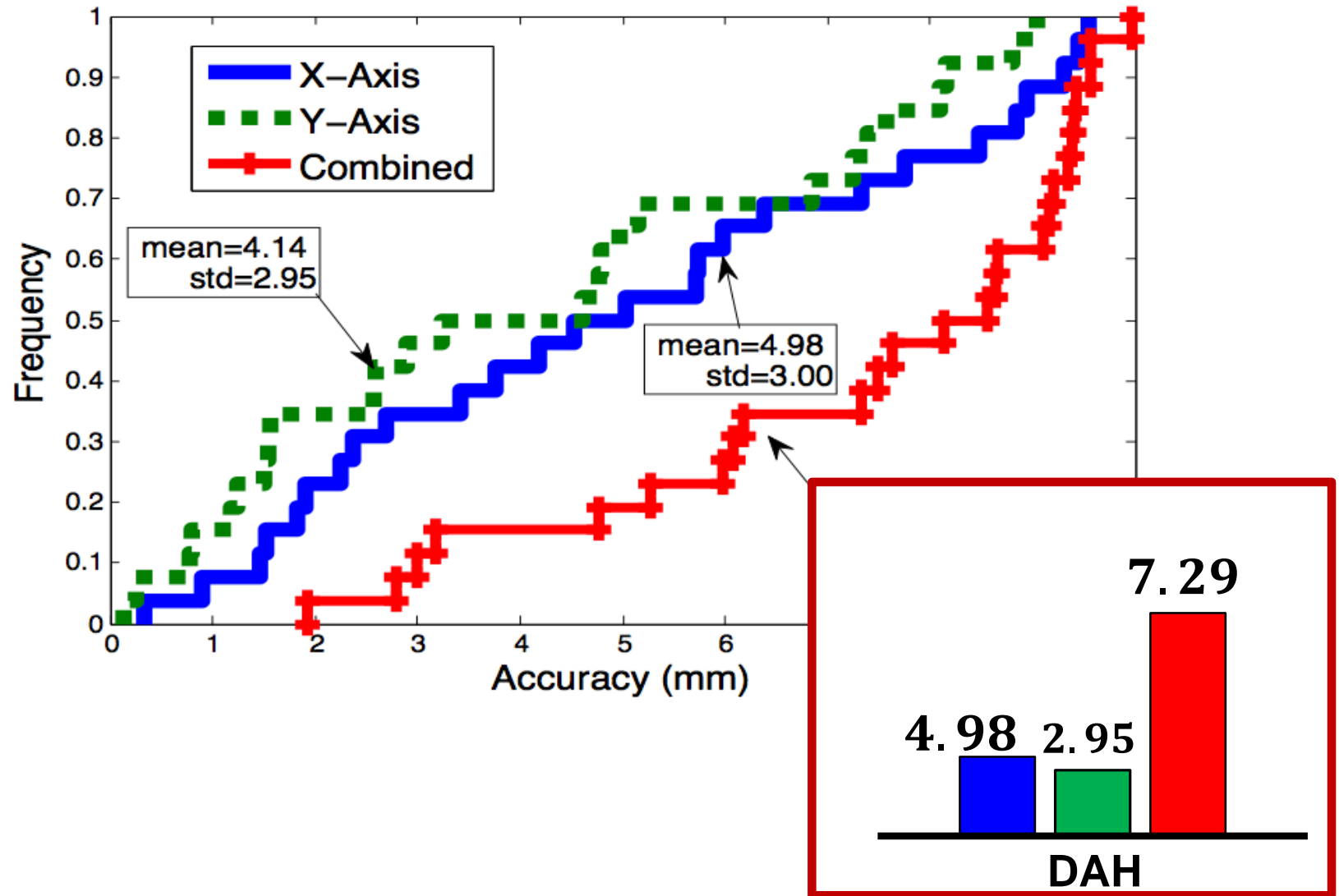


Improve the accuracy by $48 \times$, $30 \times$, $28 \times$ and $8.5 \times$ in comparison to RSS, OTrack, PinIt and BackPos.

Controllable Case with Nonlinear Track



Nonlinear track



Under Uncontrollable Case

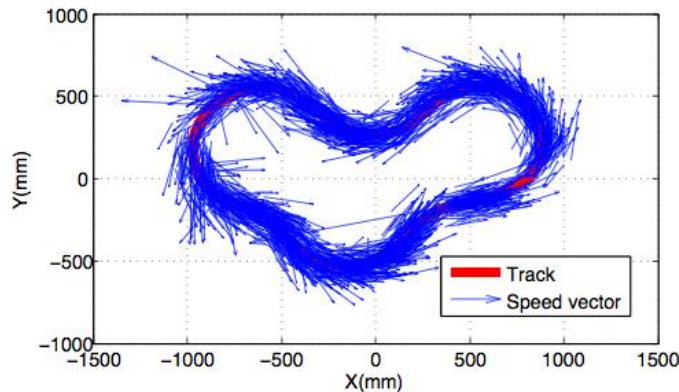
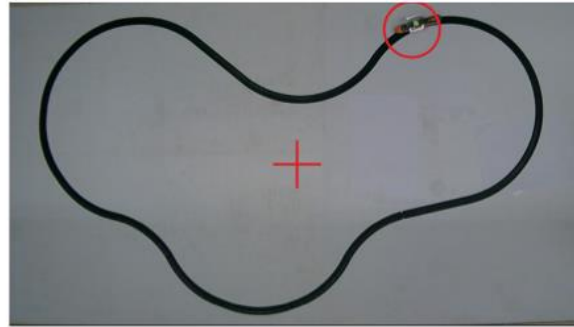


Figure 16: Estimated speed

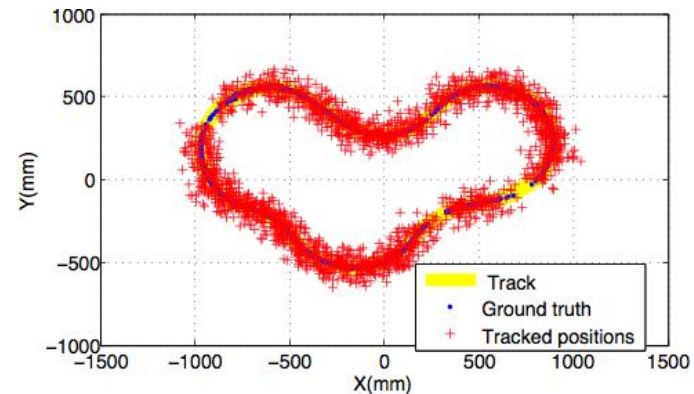


Figure 17: Fitted trajectory

Achieve a median of 12.3cm accuracy with a standard deviation of 5cm.



7

Pilot Study

In two airports

Current workflow – Manual sortation



It is error-prone step to find the baggage from the carousel in manual sortation.

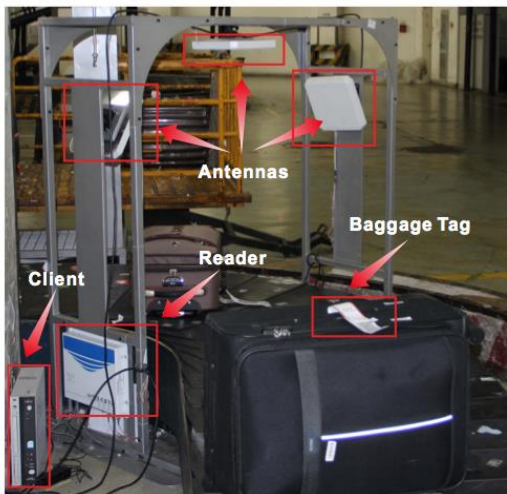


Sortation carousel



Sorting baggage

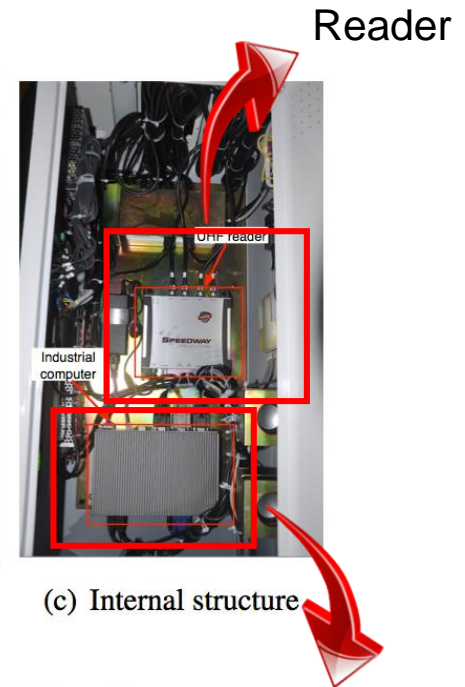
Our system: TrackPoint



(a) Version 1.0



(b) Version 2.0



(c) Internal structure



(a) Two TrackPoints



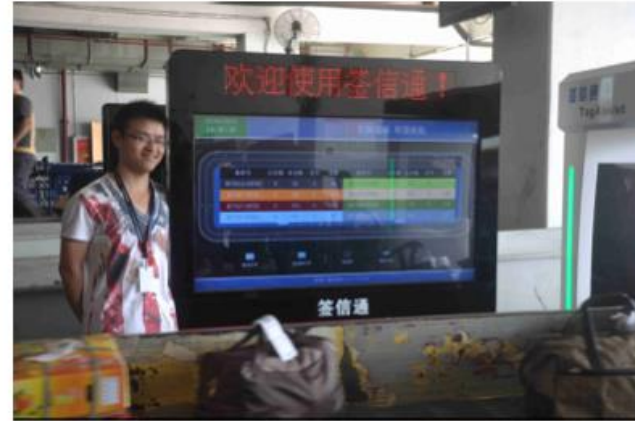
(b) Version 2.0

Industrial
computer

Tracking Visualization



(a) TrackPoint & Visualization



(b) Display screen



(c) Screenshot



(d) TrackPoint & Visualization

Pilot site



**Beijing Capital
International Airport**



**Sanya Phoenix
International Airport**

Setup

- Each airport contains **5** TrackPoints, **4** visualization screens, and **22** RFID Printers.

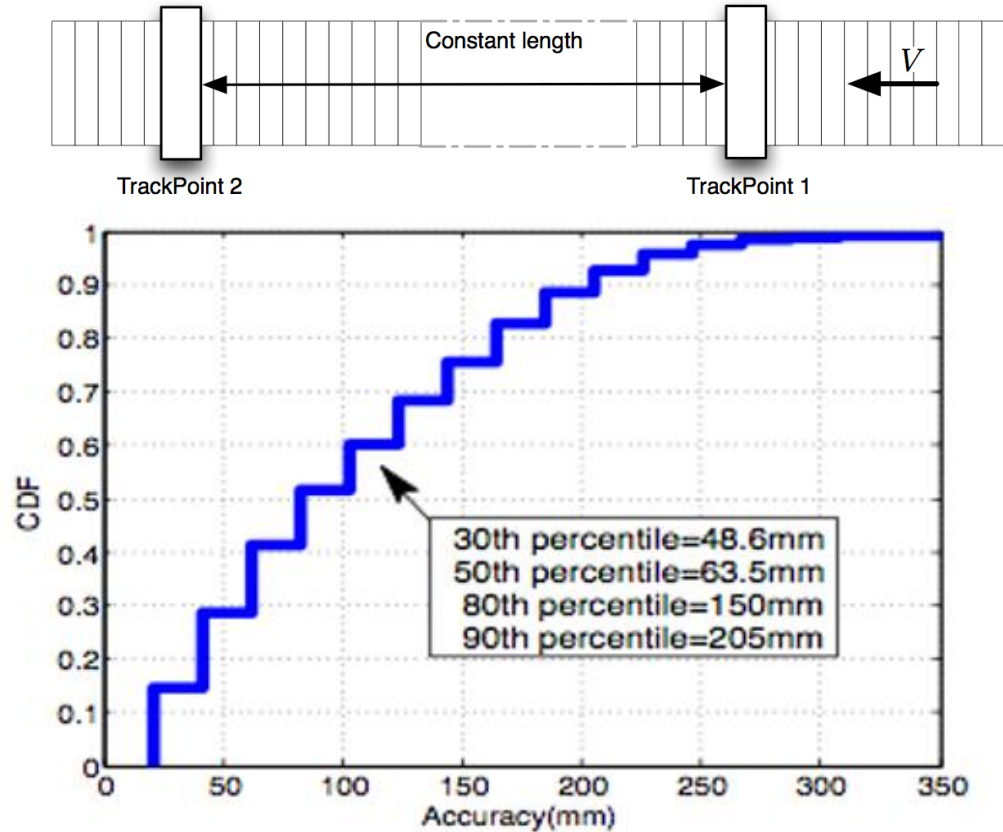
- The two-year pilot study totally spent more than **\$600,000**

Setup

- Consumed **110,000** RFID tags.

- Involved **53** destination airports, **93** air lines, and **1,094** flights.

Tracking Accuracy



Tagoram achieves a median accuracy of 6.35cm in practice.

Conclusion

- Provides **real-time** tracking of **mobile tags** with accuracy to **mm-level**.
- Limits almost all **negative impacts**.
 - Multipath effect
 - Doppler effect
 - Thermal noise
 - Device diversity

Not well-studied before
- Implemented purely based on **COTS** RFID products.
- Systematically evaluated under indoor environment and at **two airports**.



Questions?



Thank you